

AMENDMENTS TO THE CLAIMS

A1 1. (Original) A system for early warning in an e-service management system, comprising:
a statistical learning mechanism for performing statistical learning based on a plurality of data values of a variable to generate a statistical model characterizing the behavior of the variable;

an early warning mechanism for generating an early warning of threshold violation of the variable with respect to a threshold by predicting, based on the statistical model, a future time by which the values of the variable exceeds the threshold; and

an operational mechanism for detecting abnormal behavior of the variable based on both the statistical model and the early warning.

2. (Currently amended) The system according to claim 1, wherein the statistical learning mechanism comprises:

an offline normal behavior modeling mechanism for modeling the regular behavior of the variable based on the plurality of values of the variable collected offline over a period of time; and

an online behavior modeling mechanism for ~~modeling~~ modeling the dynamic behavior of the variable based on a plurality of values of the variable collected online during the operations performed by the operational mechanism.

3. (Currently amended) A method for early warning in an e-service management system, comprising:

modeling the behavior of a variable based on a plurality of data values of the variable collected over a period of time, said modeling being performed based on the statistical properties

A) of the data values of the variable to generate a behavior model for the variable, the behavior model being represented using a plurality of model parameters;

generating an early warning for a threshold violation of the variable with respect to a threshold based on a plurality of data values of the variable collected online and the behavior model; and

detecting abnormal behavior of the variable according to the plurality of data values of the variable collected online and the early warning.

4. (Original) The method according to claim 3, wherein the modeling comprises:

establishing, by an offline normal behavior modeling mechanism, a first statistical model that characterizes the regular behavior of the variable based on a first set of values of the variable collected offline over a period of time; and

establishing a second statistical model that characterizes the dynamic behavior of the variable based on a second set of values of said variable collected online, said first and said second statistical model comprising said behavior model.

5. (Original) The method according to claim 3, wherein generating an early warning comprises:

computing a plurality of residuals at corresponding different time reference points in the future based on the model parameters; deriving the variances of the plurality of residuals, predicted by said predicting;

estimating the probabilities for threshold violation of the variable with respect to said threshold at the corresponding different time reference points in the future; and

issuing an early warning for any of the time reference points at which the probability for

threshold violation of the variable exceeds a pre-determined value.

A1

6. (Currently amended) The method according to claim 5, wherein the estimating the probabilities comprises:

translating the threshold for the variable to corresponding residual threshold for the residual of the variable; and

calculating the probabilities for threshold violation of the residual with respect to the residual threshold at the corresponding different time reference points in the future.

7. (Currently amended) A computer-readable medium encoded with a program for early warning in an e-service management system, the program comprising: ~~program, when executed,~~ causing:

modeling the behavior of a variable based on a plurality of data values of the variable collected over a period of time, said modeling being performed based on the statistical properties of the data values of the variable to generate a behavior model for the variable, the behavior model being represented using a plurality of model parameters;

generating an early warning for a threshold violation of the variable with respect to a threshold based on a plurality of data values of the variable collected online and the behavior model; and

detecting abnormal behavior of the variable according to the plurality of data values of the variable collected online and the early warning.

8. (Original) The medium according to claim 7, wherein the modeling comprises:

establishing, by an offline normal behavior modeling mechanism, a first statistical model

that characterizes the regular behavior of the variable based on a first set of values of the variable collected offline over a period of time; and

A1

establishing a second statistical model that characterizes the dynamic behavior of the variable based on a second set of values of said variable collected online, said first and said second statistical model comprising said behavior model.

9. (Currently amended) The medium according to claim 7, wherein generating an early warning comprises:

computing a plurality of residuals at ~~corresponding~~ corresponding to different time reference points in the future based on the model parameters;

deriving the variances of the plurality of residuals, predicted by said predicting;

estimating the probabilities for threshold violation of the variable with respect to said threshold at the corresponding different time reference points in the future; and

issuing an early warning for any of the time reference points at which the probability for threshold violation of the variable exceeds a pre-determined value.

10. (Currently amended) The medium according to claim 9, wherein the estimating the probabilities comprises:

translating the threshold for the variable to corresponding residual threshold for the residual of the variable; and

calculating the probabilities for threshold violation of the residual with respect to the residual threshold at the corresponding different time reference points in the future.

11. (New) The system as claimed in claim 1, wherein the thresholds corresponding to future times are distinct based on time of day and day of week.

A2

12. (New) The system as claimed in claim 1, wherein the detection of abnormal behavior is further based on criteria derived from a business process model.

13. (New) The system as claimed in claim 2, wherein the online behavior modeling mechanism includes a sliding window.

14. (New) The system as claimed in claim 13, wherein the sliding window is used to determine stationarity segments of the variable and the times corresponding to boundaries of the stationarity segments.

15. (New) The system as claimed in claim 13, wherein the width of the sliding window is configurable parameter.

16. (New) The method as claimed in claim 3, wherein the modeling of the behavior of the variable includes detecting time segments of stationarity and the times corresponding to the boundaries of these stationarity segments.

17. (New) The method as claimed in claim 3, wherein detecting abnormal behavior of the variable is further according to criteria derived from a business process model.

18. (New) The method as claimed in claim 5, wherein the residuals corresponding to different time points in the future comprise a distinct value depending on the time of day, day of week, and week of month of the future time.

19. (New) The method as claimed in claim 5, wherein the number of time reference points and the pre-determined threshold violation value are configurable parameters.

20. (New) The method as claimed in claim 5, wherein the number of time reference points and the pre-determined threshold violation value are derived from a business process model.

21. (New) The computer-readable medium as claimed in claim 7, wherein the threshold is further based on criteria derived from a business process model.

22. (New) The computer-readable medium as claimed in claim 7, wherein the threshold is a distinct value based on time of day and day of week.

23. (New) The computer-readable medium as claimed in claim 8, wherein the modeling of the behavior of the variable includes detecting time segments of stationarity and the times corresponding to the boundaries of these stationarity segments.

24. (New) The computer-readable medium as claimed in claim 8, wherein modeling of the behavior of the variable includes the use of a sliding window.